DOCUMENT RESUME

ED 426 095 TM 029 325

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TITLE Partial and Part Correlation Coefficients: Formula and Score Regression Perspectives.
PUB DATE 1999-01-00
NOTE 19p.; Paper presented at the Annual Meeting of the Southwest Educational Research Association (San Antonio, TX, January 21-23, 1999).
PUB TYPE Reports - Descriptive (141) -- Speeches/Meeting Papers (150)
EDRS PRICE MF01/PC01 Plus Postage.
DESCRIPTORS *Correlation; *Regression (Statistics); Research Methodology; *Scores
IDENTIFIERS Statistical Package for the Social Sciences

ABSTRACT

Partial and part correlations are discussed as a means of statistical control. Partial and part correlation coefficients measure relationships between two variables while controlling for the influences of one or more other variables. They are statistical methods for determining whether a true correlation exists between a dependent and an independent variable while controlling for one or more other variables. This paper discusses the use and limitations of partial correlations, and presents heuristic data illustrating that computation formulas and regression analyses with latent scores yield equivalent results. Three appendixes contain an example of Statistical Package for the Social Sciences (SPSS) partial correlation computation, SPSS commands for regression analyses, and the SPSS regression analysis printout. (Contains 3 tables, 1 figure, and 10 references.) (SLD)

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Partial and Part Correlation Coefficients: Formula and Score Regression Perspectives

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TM029325

Paper presented at the annual meeting of the Southwest Educational Research Association, San Antonio, January, 1999.

Abstract

This paper discusses partial and part correlations as means of statistical control. Partial and part correlation coefficients measure relationships between two variables while controlling for the influences of one or more other variables. The paper discusses the use and limitations of partial correlations, and presents heuristic data illustrating that computation formulae and regression analyses with latent scores both yield equivalent results.

Partial and Part Correlation Coefficients: Formula and Score Regression Perspectives

Experimental validity is an important concern for researchers (Huck & Cormier, 1996). By implementing methods of experimental and statistical control, better confidence can be placed in one's results. These controls exist in order to reduce extraneous variability in research studies. Some common forms of experimental control include: random assignment, matching of subjects, use of experimental and control groups, and testing alternative hypotheses. While random assignment is considered one of the better variance controls, in most psychological research it is not feasible (Pedhazur, 1982). Statistical control is used as a mathematical means of comparing subjects when they can not be equivalent in fact (McBurney, 1994). Statistical methods are particularly needed when there is more than one independent variable. In these cases, the independent variables are not only related to the dependent variable, but are also related to each other. Researchers need to be able to determine the effects of some variables while controlling others. Partial and part correlations are statistical methods for determining whether a true correlation exists between a dependent and independent variable while controlling for one or more other variables (Waliczek, 1996).

Basic Ideas and Definitions

Partial and part correlation coefficients can be considered Pearson correlation coefficients and are related to multiple regression. In both partial and part correlation, two variables are correlated with the influence of one or more variables partialed out of (i.e., literally removed from) both or only one variable (Hinkle, 1998). For example, a

researcher analyzing the relationship between three variables such as height, weight, and age may want to know if a relationship exists when the age variable is removed. A partial correlation would correlate height and weight while removing the influence of age. A part correlation would correlate height and weight while the age variable is *only removed from one* of the variables.

In order to use partial correlation effectively, the data must be interval or ratio, comparably distributed, and a linear relationship must exist between variables (Korn, 1984; Waliczek, 1996). A partial correlation may be expressed as $r_{xy.z}$; this is interpreted as the correlation between x and y with z removed. Often, numbers are used rather than letters, such as $r_{12.3}$. A partial correlation with one variable controlled is known as a first-order correlation. More than one variable may be controlled at a time and the coefficients from such analyses are considered second-order, third-order, and so on based on how many independent variables are removed (Pedhazur, 1982). For example, the notation, $r_{12.34}$ implies that this is a second-order correlation with variables 3 and 4 being controlled. Often, a Pearson correlation is called a zero-order correlation due to no variables being controlled since there is only one independent variable.

Examples and Cautions

Partial correlations can be extremely valuable in detecting spurious correlations. For example, this occurs when two variables, such as x and y are correlated due to both being affected by z. However, once z is removed, the correlation between x and y disappears. According to Pedhazur (1982), one of the most common spurious correlations occurs when age is one of the variables. Researchers are not always able to test people of

the same age, therefore, partial correlation assists in determining whether age is contributing to a misleading conclusion. As Pedhazur (1982) noted,

> Partial correlation is not an all-purpose method of control. Its valid application is predicated on a sound theoretical model. Controlling variables without regard to the theoretical considerations about the patterns of relations among them may amount to a distortion of reality and result in misleading or meaningless results. (p. 110)

Figure 1 provides an example of two patterns of causation that are acceptable in performing a partial correlation along with two patterns that are not acceptable. When variables exist with complex patterns of causation, partial correlation should not be used (Pedhazur, 1982; Waliczek, 1996). As stated by Franzblau (1958), partial correlation should not be considered as a substitute for careful research technique and planning.

The formula to compute a first-order partial correlation is:

$$r_{xy\,z} = \frac{r_{xy} - r_{xz}r_{yz}}{\sqrt{1 - r^2_{xz}}\sqrt{1 - r^2_{yz}}}$$

Note that the partial correlation coefficient can be calculated directly from the three correlation coefficients: r_{xy}, r_{xz}, and r_{yz}. The same is true for a first-order part correlation. The formula for a first-order part correlation is:

$$r_{x(y\,z)} = \frac{r_{xy} - r_{xz}r_{yz}}{\sqrt{1 - r_{yz}^2}}$$

Tables 1 and 2 show the formula calculations for a partial correlation using a

computer spreadsheet program. Table 1 illustrates an extreme case in which the partialed variable *z* causes the correlation coefficient between variables *x* and *y* to increase. Table 2 illustrates another extreme case where the effect is the opposite.

In cases where more than one independent variable is partialed out, the same formulas for partial and part correlations are utilized. For example, if a second-order partial correlation is hand-calculated, one must first calculate three first-order partials to plug into the formula. The formula would appear as:

$$r_{12.34} = \frac{r_{12.3} - r_{14.3}r_{24.3}}{\sqrt{1 - r^2_{14.3}}\sqrt{1 - r^2_{24.3}}}.$$

Another simple approach to computing partial correlation coefficients is by using a computer statistical software program. The Statistical Package for Social Sciences (SPSS) is able to quickly calculate partial correlations. Table 3 contains a small data set and the results from SPSS analysis of these data are illustrated in Appendix A.

As stated earlier, partial and part correlations are related to regression.

> In simplest terms, the partial correlation $r_{12.3}$ is merely the correlation between the residual from predicting X_1 from X_3 and the residual from predicting X_2 from X_3. (Hays, 1994, p. 675)

In regression, the correlation between an independent variable (predictor) and residuals of another variable is always zero. In other words, error scores are always uncorrelated with yhat scores. In a first-order partial correlation with *X*, *Y*, and *Z* variables, two regression equations would be obtained. In partialing out *Z*, we would have one equation for predicting *X* from *Z* and one for *Y* from *Z*. The partial correlation is simply the

correlation of the residuals or error scores of *X* and *Y*. An SPSS example is presented here to show that computation formulae and regression analyses with yhat and "e" scores both yield equivalent results.

Using the small data set from Table 3, two regression analyses were performed. Appendix B contains the SPSS commands used in this example. Data for b weights and y-intercepts were obtained and entered into prediction equations. This resulted in a final computer printout, Appendix C, that contains the correlation coefficient between the *X* "e" scores and the *Y* "e" scores. By comparing the score regression analyses to the formula computation in Appendix A, we find that the values are identical.

Summary

Partial and part correlations can be usefully applied in many research situations. In fact, statistically controlling for an extraneous variable may even result in a part or partial correlation that is closer to +1 or to –1 (i.e., further from zero) than the original r, if "suppressor effects" are involved (see Henard, 1998).

However, statistical control mechanisms such as these must be used thoughtfully (Franzblau, 1958). A particular problem that can occur, particularly if the control involves quite a few orders (e.g., second-order, third-order, fourth-order) is that it may no longer be clear what is being correlated, after removing so much variance from the two measured variables of primary interest (Thompson, 1995).

References

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Table 1: Partial Correlation Computation

r_{xy}	r_{xz}	r_{yz}	r_{xz}^2	r_{yz}^2
0.6	0.8	0.1	0.64	0.01

$$\text{Partial1} = \frac{r_{xy} - r_{xz}r_{yz}}{\text{sqr}\,(1-r_{xz}^2)(1-r_{yz}^2)}$$

$$= \frac{0.6-0.8(0.1)}{\text{sqr}(0.36)(0.99)}$$

$$= \frac{0.6-0.8}{\text{sqr}(0.3564)}$$

$$= \frac{0.52}{0.596992}$$

$$= 0.871033$$

Table 2: Partial Correlation Computation

r_{xy}	r_{xz}	r_{yz}	r_{xz}^2	r_{yz}^2
0.75	0.8	0.9	0.64	0.81

$$\text{Partial2} = \frac{r_{xy} - r_{xz}r_{yz}}{\text{sqr}\,(1-r_{xz}^2)(1-r_{yz}^2)}$$

$$= \frac{0.75-0.8(0.9)}{\text{sqr}(0.36)(0.19)}$$

$$= \frac{0.75-0.72}{\text{sqr}(0.0684)}$$

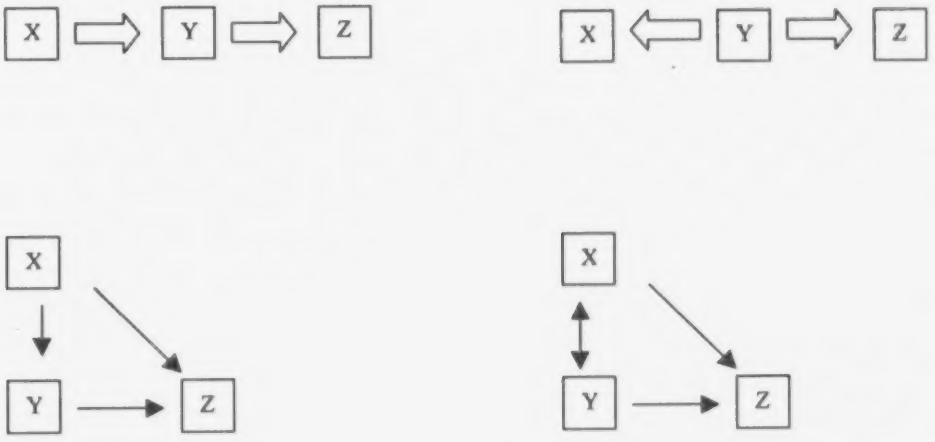
$$= \frac{0.03}{0.261534}$$

$$= 0.114708$$

Table 3: Example Data Set

ID	X	Y	Z
1	15	48	10
2	20	68	34
3	25	36	39
4	30	42	46
5	35	14	17
6	40	60	22
7	45	47	56
8	50	72	60
9	55	80	65
10	60	91	59

Figure 1: Models of Causation



Appendix A

Example SPSS Partial Correlation Computation

```
PARTIAL CORR
    /VARIABLES= var00001 var00002 BY var00003
  /SIGNIFICANCE=TWOTAIL
  /MISSING=LISTWISE .

- -P A R T I A L   C O R R E L A T I O N   C O E F F I C I E N T S - -

Controlling for.. VAR Z

                VAR X       VAR Y

VAR X          1.0000       .2256
               (    0)     (    7)
               P= .        P= .559

VAR Y           .2256      1.0000
               (    7)     (    0)
               P= .559     P= .

(Coefficient / (D.F.) / 2-tailed Significance)
```

Appendix B

SPSS Commands for Regression Analyses

```
TITLE 'Partial correlation'.
COMMENT**************************************.
COMMENT SPSS example.
COMMENT SERA.
COMMENT**************************************.
SET BLANKS=SYSMIS UNDEFINED=WARN PRINTBACK=LISTING.
DATA LIST
     FILE='A:\partial.txt' FIXED RECORDS=1
     /ID 1-2 V1 4-5 V2 7-8 V3 10-11.
EXECUTE.
CORRELATIONS
  /VARIABLES=v1 v2 v3
  /PRINT=TWOTAIL NOSIG
  /MISSING=PAIRWISE .
REGRESSION
  /MISSING LISTWISE
  /STATISTICS COEFF OUTS R ANOVA
  /CRITERIA=PIN(.05) POUT(.10)
  /NOORIGIN
  /DEPENDENT v1
  /METHOD=ENTER v3 .
REGRESSION
  /MISSING LISTWISE
  /STATISTICS COEFF OUTS R ANOVA
  /CRITERIA=PIN(.05) POUT(.10)
  /NOORIGIN
  /DEPENDENT v2
  /METHOD=ENTER v3 .

compute yhat1=13.986+ (.576*V3) .
compute e1=V1-yhat1 .
compute yhat2=27.034+ (.705*V3) .
compute e2=V2-yhat2 .
LIST VARIABLES=YHAT1 E1/CASES=99 .
LIST VARIABLES=YHAT2 E2/CASES=99 .
correlations variables=V1 V2 V3 yhat1 yhat2 e1 e2/STATISTICS=ALL .
PARTIAL CORR VARIABLES=V1 WITH V2 BY V3 (1) .
```

APPENDIX C

SPSS Regression Analysis Printout

```
09 Nov 98   Partial correlation                                    Page 1

-> DATA LIST
->      FILE='A:\partial.txt' FIXED RECORDS=1
->      /ID 1-2 V1 4-5 V2 7-8 V3 10-11.
-> EXECUTE.
->
-> CORRELATIONS
->   /VARIABLES=v1 v2 v3
->   /PRINT=TWOTAIL NOSIG
->   /MISSING=PAIRWISE .

09 Nov 98   Partial correlation                                    Page 2

                          - -  Correlation Coefficients  - -

             V1          V2          V3

V1          1.0000       .5704       .7466*
V2           .5704      1.0000       .6037
V3           .7466*      .6037      1.0000

* - Signif. LE .05      ** - Signif. LE .01      (2-tailed)          " . " is
printed if a coefficient cannot be computed

09 Nov 98   Partial correlation                                    Page 3
-> REGRESSION
->   /MISSING LISTWISE
->   /STATISTICS COEFF OUTS R ANOVA
->   /CRITERIA=PIN(.05) POUT(.10)
->   /NOORIGIN
->   /DEPENDENT v1
->   /METHOD=ENTER v3 .

09 Nov 98   Partial correlation                                    Page 4

                              * * * *  M U L T I P L E   R E G R E S S I O N
 * * * *

Listwise Deletion of Missing Data

Equation Number 1    Dependent Variable..   V1

Block Number  1.  Method:  Enter      V3

Variable(s) Entered on Step Number  1..    V3

Multiple R            .74663        Analysis of Variance
R Square              .55746                    DF      Sum of Squares     Mean Square
Adjusted R Square     .50214        Regression   1          1149.76456       1149.76456
Standard Error      10.68138        Residual     8           912.73544        114.09193

                                        F =       10.07753       Signif F =  .0131
------------------ Variables in the Equation ------------------

Variable              B         SE B       Beta          T  Sig T
```

```
V3              .576323     .181547    .746634     3.175  .0131
(Constant)    13.986018    8.140920                1.718  .1241

End Block Number   1   All requested variables entered.

09 Nov 98   Partial correlation                                    Page 5
-> REGRESSION
->   /MISSING LISTWISE
->   /STATISTICS COEFF OUTS R ANOVA
->   /CRITERIA=PIN(.05) POUT(.10)
->   /NOORIGIN
->   /DEPENDENT v2
->   /METHOD=ENTER v3 .

09 Nov 98   Partial correlation                                    Page 6

                            * * * *  M U L T I P L E  R E G R E S S I O N
 * * * *

Listwise Deletion of Missing Data

Equation Number 1    Dependent Variable..   V2

Block Number  1.  Method:  Enter      V3

Variable(s) Entered on Step Number  1..     V3

Multiple R          .60369        Analysis of Variance
R Square            .36444                    DF      Sum of Squares      Mean Square
Adjusted R Square   .28500        Regression   1         1720.74427        1720.74427
Standard Error    19.36768        Residual     8         3000.85573         375.10697

                                        F =       4.58734       Signif F =  .0646
------------------ Variables in the Equation ------------------

Variable               B         SE B        Beta          T  Sig T

V3               .705050     .329184    .603689     2.142  .0646
(Constant)     27.033973   14.761264                1.831  .1044

End Block Number   1   All requested variables entered.

09 Nov 98   Partial correlation                                    Page 7
-> compute yhat1=13.986+ (.576*V3) .
-> compute e1=V1-yhat1 .
-> compute yhat2=27.034+ (.705*V3) .
-> compute e2=V2-yhat2 .
-> LIST VARIABLES=YHAT1 E1/CASES=99 .

09 Nov 98   Partial correlation                                    Page 8

   YHAT1        E1

   19.75     -4.75
   33.57    -13.57
   36.45    -11.45
   40.48    -10.48
   23.78     11.22
   26.66     13.34
   46.24     -1.24
   48.55      1.45
```

```
   51.43     3.57
   47.97    12.03

Number of cases read:  10    Number of cases listed:  10

09 Nov 98   Partial correlation                                    Page 9
-> LIST VARIABLES=YHAT2 E2/CASES=99 .

09 Nov 98   Partial correlation                                   Page 10

   YHAT2       E2

   34.08    13.92
   51.00    17.00
   54.53   -18.53
   59.46   -17.46
   39.02   -25.02
   42.54    17.46
   66.51   -19.51
   69.33     2.67
   72.86     7.14
   68.63    22.37

Number of cases read:  10    Number of cases listed:  10

09 Nov 98   Partial correlation                                   Page 11
-> correlations variables=V1 V2 V3 yhat1 yhat2 e1 e2/STATISTICS=ALL .

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Variable     Cases          Mean          Std Dev

V1              10       37.5000          15.1383
V2              10       55.8000          22.9046
V3              10       40.8000          19.6118
YHAT1           10       37.4868          11.2964
YHAT2           10       55.7980          13.8263
E1              10         .0132          10.0705
E2              10         .0020          18.2600

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Variables            Cases   Cross-Prod Dev   Variance-Covar

V1       V2             10        1780.0000         197.7778
V1       V3             10        1995.0000         221.6667
V1       YHAT1          10        1149.1200         127.6800
V1       YHAT2          10        1406.4750         156.2750
V1       E1             10         913.3800         101.4867
V1       E2             10         373.5250          41.5028
V2       V3             10        2440.6000         271.1778
V2       YHAT1          10        1405.7856         156.1984
V2       YHAT2          10        1720.6230         191.1803
V2       E1             10         374.2144          41.5794
V2       E2             10        3000.9770         333.4419
V3       YHAT1          10        1993.8816         221.5424
V3       YHAT2          10        2440.4280         271.1587
V3       E1             10           1.1184            .1243
V3       E2             10            .1720            .0191
YHAT1    YHAT2          10        1405.6865         156.1874
YHAT1    E1             10            .6442            .0716
```

YHAT1	E2	10	.0991	.0110
YHAT2	E1	10	.7885	.0876
YHAT2	E2	10	.1213	.0135
E1	E2	10	373.4259	41.4918

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- - Correlation Coefficients - -

	V1	V2	V3	YHAT1	YHAT2	E1	E2
V1	1.0000	.5704	.7466	.7466	.7466	.6657	.1501
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .	P= .085	P= .013	P= .013	P= .013	P= .036	P= .679
V2	.5704	1.0000	.6037	.6037	.6037	.1803	.7973
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .085	P= .	P= .065	P= .065	P= .065	P= .618	P= .006
V3	.7466	.6037	1.0000	1.0000	1.0000	.0006	.0001
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .013	P= .065	P= .	P= .000	P= .000	P= .999	P=1.000
YHAT1	.7466	.6037	1.0000	1.0000	1.0000	.0006	.0001
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .013	P= .065	P= .000	P= .	P= .000	P= .999	P=1.000
YHAT2	.7466	.6037	1.0000	1.0000	1.0000	.0006	.0001
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .013	P= .065	P= .000	P= .000	P= .	P= .999	P=1.000
E1	.6657	.1803	.0006	.0006	.0006	1.0000	.2256
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .036	P= .618	P= .999	P= .999	P= .999	P= .	P= .531
E2	.1501	.7973	.0001	.0001	.0001	.2256	1.0000
	(10)	(10)	(10)	(10)	(10)	(10)	(10)
	P= .679	P= .006	P=1.000	P=1.000	P=1.000	P= .531	P= .

(Coefficient / (Cases) / 2-tailed Significance) " . " is printed if a coefficient cannot be computed

09 Nov 98 Partial correlation Page 15
-> PARTIAL CORR VARIABLES=V1 WITH V2 BY V3 (1) .

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- - - - - - - - - - - - - - - P A R T I A L C O R R E L A T I O N C O E F F I C I E N T S - - - - - - - - - - - - - -

Controlling for.. V3

	V2
V1	.2256
	(7)
	P= .559

(Coefficient / (D.F.) / 2-tailed Significance) " . " is printed if a coefficient cannot be computed

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